

KRZYSZTOF ZAJKOWSKI

Covariance matrices of self-affine measures

Abstract

In this paper we derive a formula for a covariance matrix of any self-affine measure, i.e. a probability measure μ satisfying

$$\mu = \sum_{k=1}^l p_k \mu \circ S_k^{-1},$$

where $\{S_k(\mathbf{x}) = A_k \mathbf{x} + \mathbf{b}_k\}_{1 \leq k \leq l}$ is a family of affine contractive maps and $\{p_k\}_{1 \leq k \leq l}$ is a set of probability weights. In particular if for every k , $A_k = A$ then the formula will be have the following form

$$D^2 X = [I \otimes I - A \otimes A]^{-1} \mathcal{D}^2 \mathcal{B},$$

where $D^2 X$ denote the covariance matrix of the measure μ and $\mathcal{D}^2 \mathcal{B}$ denote a covariance matrix of a discret random variable \mathcal{B} with values \mathbf{b}_k , and corresponding probabilities p_k .

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1 Introduction

We will consider a invariant probability measure on \mathbb{R}^d

$$(1) \quad \mu = \sum_{k=1}^l p_k \mu \circ S_k^{-1},$$

where $\{S_k\}_{1 \leq k \leq l}$ is a family of contractive maps and $\{p_k\}_{1 \leq k \leq l}$ is a set of probability weights. It is often assumed that the maps are similitudes. We will make a more general assumption that the maps are affine contractions on \mathbb{R}^d ; i.e. $S_k(\mathbf{x}) = A_k \mathbf{x} + \mathbf{b}_k$ and the operator norm $\|A_k\| < 1$ for all k .

The definition (1) was introduced by Hutchinson [H]. But an example of such measures has been studied for a long time in the context of Bernoulli convolution, i.e. the example of an invariant measure on the real line

$$(2) \quad \mu = \frac{1}{2}(\mu \circ S_1^{-1} + \mu \circ S_2^{-1}),$$

where $S_1(x) = \beta(x + 1)$, $S_2(x) = \beta(x - 1)$ for $\beta \in (0, 1)$. It remains difficult open problem to characterize the set of β for which μ is absolutely continuous. An another example has been studied in great detail in wavelet theory in connection with the dilatation equation

$$(3) \quad f(x) = \sum_{k=1}^l c_k f(2x - (k - 1)).$$

The function f can be considered as the density function of the corresponding absolutely continuous self-affine measure μ for $S_k(x) = \frac{1}{2}(x + (k - 1))$ and $p_k = \frac{1}{2}c_k$. In wavelet theory the c_k may be negative but $\sum c_k$ must be 2. The invariant measures have a natural connection with fractal geometry [F], because their supports are compact invariant sets

$$(4) \quad K = \bigcup_{k=1}^l S_k(K).$$

These measures arise also in another areas of mathematics.

A fundamental method for studing these measures is the Fourier transform [S]. Our goal is to show that we can use another probabilistic tools, not only characteristic function, to investigate the self-affine measures. We derive a formula for covariance matrices and give an example of an investigation of measures on Sierpinski triangle.

2 Covariance matrices of self-affine measures

The invariant measure μ satisfies the following identity

$$(5) \quad \int_{\mathbb{R}^d} f d\mu = \sum_{k=1}^l p_k \int_{\mathbb{R}^d} f \circ S_k d\mu,$$

where f is any continuous function on \mathbb{R}^d [B]. We will denote by X some d -dimensional random variable with respect to the probability distribution μ . We apply the identity (5) to the coordinate functions $e_i^*(\mathbf{x}) = x_i$ of the point $\mathbf{x} = (x_i)_{1 \leq i \leq d} \in \mathbb{R}^d$. By the above a vector of expected values EX will be equal

$$\begin{aligned}
EX &= \left(\int_{\mathbb{R}^d} e_i^*(\mathbf{x}) d\mu \right)_{1 \leq i \leq d} = \left(\sum_{k=1}^l p_k \int_{\mathbb{R}^d} e_i^*(A_k \mathbf{x} + \mathbf{b}_k) d\mu \right)_{1 \leq i \leq d} \\
(6) \quad &= \sum_{k=1}^l p_k E(A_k X + \mathbf{b}_k) = \sum_{k=1}^l p_k A_k EX + \sum_{k=1}^l p_k \mathbf{b}_k
\end{aligned}$$

The above relation gives

$$(7) \quad \left[I - \sum_{k=1}^l p_k A_k \right] EX = \sum_{k=1}^l p_k \mathbf{b}_k,$$

where I is identity matrix on \mathbb{R}^d . The sum $\sum_{k=1}^l p_k \mathbf{b}_k$ is a vector of expected values of a d -dimensional random variable \mathcal{B} with values \mathbf{b}_k , and corresponding probabilities p_k . Since $\|A_k\| < 1$ for all k then $\|\sum_{k=1}^l p_k A_k\| < 1$. It follows that 1 is not an eigenvalue of the operator $\sum_{k=1}^l p_k A_k$. For this reason the operator $I - \sum_{k=1}^l p_k A_k$ will be invertible. We can rewrite (7) as

$$(8) \quad EX = \left[I - \sum_{k=1}^l p_k A_k \right]^{-1} \mathcal{E} \mathcal{B}.$$

This means that the expected value of X linearly depend on the expected value of \mathcal{B} .

Let $X \otimes X = [x_i x_j]_{1 \leq i, j \leq d}$ denote the second order tensor build from the coordinates. Using (5) a matrix of second order moments $E(X \otimes X)$ will be equal

$$\begin{aligned}
E(X \otimes X) &= \left[\int_{\mathbb{R}^d} e_i^*(\mathbf{x}) e_j^*(\mathbf{x}) d\mu \right]_{1 \leq i, j \leq d} \\
&= \sum_{k=1}^l p_k \left[\int_{\mathbb{R}^d} e_i^*(A_k \mathbf{x} + \mathbf{b}_k) e_j^*(A_k \mathbf{x} + \mathbf{b}_k) d\mu \right]_{1 \leq i, j \leq d} \\
&= \sum_{k=1}^l p_k E((A_k X + \mathbf{b}_k) \otimes (A_k X + \mathbf{b}_k)) \\
(9) \quad &= \sum_{k=1}^l p_k [(A_k \otimes A_k) E(X \otimes X) + \mathbf{b}_k \otimes A_k EX + A_k EX \otimes \mathbf{b}_k + \mathbf{b}_k \otimes \mathbf{b}_k]
\end{aligned}$$

Therefore

$$[I \otimes I - \sum_{k=1}^l p_k(A_k \otimes A_k)]E(X \otimes X) = \sum_{k=1}^l p_k(\mathbf{b}_k \otimes A_k EX + A_k EX \otimes \mathbf{b}_k + \mathbf{b}_k \otimes \mathbf{b}_k) \quad (10)$$

The operator norm of $A_k \otimes A_k$ is less than 1 on $\mathbb{R}^d \otimes \mathbb{R}^d$, so, by the same argument as early, we get that the operator $I \otimes I - \sum_{k=1}^l p_k(A_k \otimes A_k)$ is invertible and we obtain

$$E(X \otimes X) = [I \otimes I - \sum_{k=1}^l p_k(A_k \otimes A_k)]^{-1} \times \sum_{k=1}^l p_k(\mathbf{b}_k \otimes A_k EX + A_k EX \otimes \mathbf{b}_k + \mathbf{b}_k \otimes \mathbf{b}_k). \quad (11)$$

Substituting (8) into $D^2X = E(X \otimes X) - EX \otimes EX$ we can obtain a formula for the covariance matrix of X . But in the general case it will be complicated. This formula takes a surprising simple form when we assumed that all affine maps have the same linear part, i.e. all $A_k = A$. Under this assumption, using the standard tensor calculus we get

$$\begin{aligned} D^2X &= [I \otimes I - A \otimes A]^{-1} \{ [I \otimes A(I - A)^{-1}](\mathcal{E}\mathcal{B} \otimes \mathcal{E}\mathcal{B}) \\ &\quad + [A(I - A)^{-1} \otimes I](\mathcal{E}\mathcal{B} \otimes \mathcal{E}\mathcal{B}) + \mathcal{E}(\mathcal{B} \otimes \mathcal{B}) \} \\ &\quad - [(I - A)^{-1} \otimes (I - A)^{-1}](\mathcal{E}\mathcal{B} \otimes \mathcal{E}\mathcal{B}) \\ (12) \quad &= [I \otimes I - A \otimes A]^{-1} (\mathcal{E}(\mathcal{B} \otimes \mathcal{B}) - \mathcal{E}\mathcal{B} \otimes \mathcal{E}\mathcal{B}) \end{aligned}$$

Notice now that the expression $\mathcal{E}(\mathcal{B} \otimes \mathcal{B}) - \mathcal{E}\mathcal{B} \otimes \mathcal{E}\mathcal{B}$ is the covariance matrix of \mathcal{B} . Thus we obtain

$$(13) \quad D^2X = [I \otimes I - A \otimes A]^{-1} \mathcal{D}^2\mathcal{B}.$$

In other words we obtained the following proposition.

PROPOSITION

Assume that μ is a self-affine measure on \mathbb{R}^d for a family of linear contractions $S_k(\mathbf{x}) = A\mathbf{x} + \mathbf{b}_k$, $1 \leq k \leq l$. Let X be some random variable with respect to the probability distribution μ . Then the covariance matrix of the random variable X

$$(14) \quad D^2X = [I \otimes I - A \otimes A]^{-1} \mathcal{D}^2\mathcal{B},$$

where $\mathcal{D}^2\mathcal{B}$ denote the covariance matrix of the random variable \mathcal{B} . \square

Remark. When all affine maps S_k have the same linear part then not only the expected value of X linearly depend on expected value of \mathcal{B} but also covariance matrix of X linearly depend on the covariance matrix of \mathcal{B} .

If the matrix A is diagonal then diagonal is the matrix $[I \otimes I - A \otimes A]^{-1}$. Therefore we get the simple corollary.

COROLLARY

If under the assumptions of Proposition we assume additionally that the matrix A is diagonal then $X_i = e_i^*(X)$ and $X_j = e_j^*(X)$ are uncorrelated if and only if uncorrelated are $(e_i^*(\mathbf{b}_k))_{1 \leq k \leq l}$ and $(e_j^*(\mathbf{b}_k))_{1 \leq k \leq l}$. \square

In other words we have obtained the following law

$$\int_{\mathbb{R}^d} x_i x_j d\mu = \int_{\mathbb{R}^d} x_i d\mu \int_{\mathbb{R}^d} x_j d\mu \quad \text{iff} \quad \sum_{k=1}^l p_k e_i^*(\mathbf{b}_k) e_j^*(\mathbf{b}_k) = \sum_{k=1}^l p_k e_i^*(\mathbf{b}_k) \sum_{k=1}^l p_k e_j^*(\mathbf{b}_k).$$

3 Example

Consider Sierpinski triangle with vertices at the points $(0,0)$, $(1,0)$ and $(\frac{1}{2}, \frac{\sqrt{3}}{2})$. The Sierpinski triangle is an invariant compact set of three contractions on \mathbb{R}^2 : $S_1(\mathbf{x}) = \frac{1}{2}\mathbf{x}$, $S_2(\mathbf{x}) = \frac{1}{2}\mathbf{x} + (\frac{1}{2}, 0)$ and $S_3(\mathbf{x}) = \frac{1}{2}\mathbf{x} + (\frac{1}{4}, \frac{\sqrt{3}}{4})$. In this case the matrix $A = \frac{1}{2}I$. Let μ denote an invariant measure for weights p_1 , p_2 and p_3 . The expected value $\mathcal{E}\mathcal{B} = (\frac{1}{2}p_2 + \frac{1}{4}p_3, \frac{\sqrt{3}}{4}p_3)$. The matrix $[I - A]^{-1} = 2I$. By the (8) we get $\int_{\mathbb{R}^2} x_1 d\mu = p_2 + \frac{1}{2}p_3$ and $\int_{\mathbb{R}^2} x_2 d\mu = \frac{\sqrt{3}}{2}p_3$. By the (14) we can obtain terms of the matrix $\mathcal{D}^2\mathcal{B}$ and D^2X . In particular

$$(15) \quad \sum_{k=1}^3 p_k e_1^*(\mathbf{b}_k) e_2^*(\mathbf{b}_k) - \sum_{k=1}^3 p_k e_1^*(\mathbf{b}_k) \sum_{k=1}^3 p_k e_2^*(\mathbf{b}_k) = \frac{\sqrt{3}}{16} p_3 (p_1 - p_2).$$

By corollary, if $p_1 = p_2$ then the random variables X_1 , X_2 are uncorrelated and

$$(16) \quad \begin{aligned} \int_{\mathbb{R}^2} x_1 x_2 d\mu &= \int_{\mathbb{R}^2} x_1 d\mu \int_{\mathbb{R}^2} x_2 d\mu \\ &= (p_2 + \frac{1}{2}p_3) \frac{\sqrt{3}}{2} p_3 = \frac{\sqrt{3}}{4} p_3. \end{aligned}$$

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Krzysztof Zajkowski
 Institute of Mathematics, University of Białystok
 Akademicka 2, 15-267 Białystok, Poland
 E-mail:kryza@math.uwb.edu.pl